

Econometrics I: Solutions of the homework #6

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Contents

1 Question 1	1
1.1 (1)	1
1.2 (2)	2
1.3 (3)	3
1.4 (4)	4
1.5 (5)	5
2 Question 2	6
2.1 (6)	6
2.2 (7)	6
2.3 (8)	7

1 Question 1

1.1 (1)

In order to prove:

$$\frac{(\hat{\beta} - \beta)' X' X (\hat{\beta} - \beta)}{\sigma^2} \sim \chi^2(k)$$

*If you have any errors in handouts and materials, please contact me via lvang12@hotmail.com

First recall the OLS estimator $\hat{\beta} = \beta + (X'X)^{-1}Xu$. Substitute $\hat{\beta} - \beta$ with $(X'X)^{-1}Xu$ then the original expression becomes:

$$\begin{aligned} (\hat{\beta} - \beta)'X'X(\beta - \hat{\beta}) &= ((X'X)^{-1}X'u)'X'X(X'X)^{-1}X'u \\ &= u'X(X'X)^{-1}X'X(X'X)^{-1}X'u = u'X(X'X)^{-1}X'u \end{aligned}$$

————— Notice —————

if $X \sim N(0, I_k)$ then $X'AX \sim \chi^2(\text{tr}(A))$ where A is symmetric and idempotent, i.e., $A'A = A$

In our case X refers to $\frac{u}{\sigma}$, A refers to $X(X'X)^{-1}X$, then we can derive:

$$\frac{u'X(X'X)^{-1}X'u}{\sigma^2} \sim \chi^2(\text{tr}(X(X'X)^{-1}X'))$$

According to trace property we know that $\text{tr}(ABC) = \text{tr}(BCA)$

$$\text{tr}(X(X'X)^{-1}X') = \text{tr}((X'X^{-1}X'X)) = \text{tr}(I_k) = k$$

Thus we proved that:

$$\frac{(\hat{\beta} - \beta)'X'X(\beta - \hat{\beta})}{\sigma^2} = \frac{u'X(X'X)^{-1}X'u}{\sigma^2} \sim \chi^2(k)$$

1.2 (2)

If random variables X and Y both follow normal distribution and $\text{Cov}(X, Y) = 0$, that indicates X and Y are independent of each other.

————— Notice ———

$$\text{Cov}(X, Y) = 0 \iff \text{Cov}(g(X), h(Y)) = 0$$

In our question $s^2 = f(e) = \frac{1}{T-k}e'e$. Therefore, to prove $\hat{\beta}$ is independent of s^2 we only need to prove $\text{Cov}(\hat{\beta}, e) = 0$

Proof:

$$\begin{aligned} \text{Cov}(e, \hat{\beta}) &= E(e(\hat{\beta} - \beta)') = E((I_T - X(X'X)^{-1}X')u((X'X)^{-1}X'u)') \\ &= E((I_T - X(X'X)^{-1}X')uu'X(X'X)^{-1}) = (I_T - X(X'X)^{-1}X')E(uu')X(X'X)^{-1} \\ &= (I_T - X(X'X)^{-1}X')(\sigma^2 I_T)X(X'X)^{-1} = \sigma^2(I_T - X(X'X)^{-1}X')X(X'X)^{-1} \\ &= \sigma^2(X(X'X)^{-1} - X(X'X)^{-1}X'X(X'X)^{-1}) = \sigma^2(X(X'X)^{-1} - X(X'X)^{-1}) = 0 \end{aligned}$$

where $e = M_x u = (I_T - P_x)u = (I_n - X(X'X)^{-1}X')u$

1.3 (3)

————— Notice ———

$$\frac{U/n}{V/m} \sim F(n, m) \text{ when } U \sim \chi^2(n), V \sim \chi^2(m), U \text{ is independent of } V$$

Previously we already proved that:

$$\begin{aligned} \frac{(\hat{\beta} - \beta)'X'X(\hat{\beta} - \beta)}{\sigma^2} &= \frac{u'X(X'X)^{-1}X'u}{\sigma^2} \sim \chi^2(k) \\ \frac{(y - X\hat{\beta})'(y - X\hat{\beta})}{\sigma^2} &\sim \chi^2(T - k) \end{aligned}$$

Thus we can obtain:

$$\frac{\frac{(\hat{\beta} - \beta)'X'X(\hat{\beta} - \beta)}{\sigma^2}/k}{\frac{(y - X\hat{\beta})'(y - X\hat{\beta})}{\sigma^2}/(T-k)} \sim F(k, T-k)$$

1.4 (4)

In order to prove this question we first rewrite the sum as the vector form:

$$\sum_{i=1}^T (y_i - \bar{y})^2 = (y - \bar{y})'(y - \bar{y})$$

where:

$$(y - \bar{y}) = \begin{pmatrix} y_1 - \bar{y} \\ y_2 - \bar{y} \\ \vdots \\ y_T - \bar{y} \end{pmatrix} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_T \end{pmatrix} - \begin{pmatrix} \bar{y} \\ \bar{y} \\ \vdots \\ \bar{y} \end{pmatrix} = y - \frac{1}{T}ii'y = (I_T - \frac{1}{T}ii')y$$

Therefore $(y - \bar{y})'(y - \bar{y})$ can be written as:

$$(y - \bar{y})'(y - \bar{y}) = y'(I_T - \frac{1}{T}ii')(I_T - \frac{1}{T}ii')y = y'(I_T - \frac{1}{T}ii')y$$

Here $I_T - \frac{1}{T}ii'$ is a symmetric and idempotent matrix and we will prove this property in the next question.

1.5 (5)

Notice that $i = (1, \dots, 1)'$, thus we can write ii' as:

$$ii' = \begin{pmatrix} 1_{1,1} & \cdots & 1_{1,T} \\ \vdots & \ddots & \vdots \\ 1_{T,1} & \cdots & 1_{T,T} \end{pmatrix}$$

then

$$I_T - \frac{1}{n}ii' = \begin{pmatrix} 1 - \frac{1}{T} & -\frac{1}{T} & \cdots & -\frac{1}{T} \\ \vdots & \ddots & & \vdots \\ -\frac{1}{T} & \cdots & 1 - \frac{1}{T} & \end{pmatrix}$$

Apparently this is a symmetric matrix i.e. $I_T - \frac{1}{n}ii' = (I_T - \frac{1}{n}ii')^T$

Next we prove idempotent property by showing $(I_T - \frac{1}{T}ii')(I_T - \frac{1}{T}ii') = I_T - \frac{1}{T}ii'$

First we expand the expression as:

$$(I_T - \frac{1}{T}ii')(I_T - \frac{1}{T}ii') = I_T - \frac{1}{T}ii' - \frac{1}{T}ii' + \frac{1}{T^2}ii'ii' \quad (1)$$

where

$$ii'ii' = \begin{pmatrix} 1_{1,1} & \cdots & 1_{1,T} \\ \vdots & \ddots & \vdots \\ 1_{T,1} & \cdots & 1_{T,T} \end{pmatrix} \begin{pmatrix} 1_{1,1} & \cdots & 1_{1,T} \\ \vdots & \ddots & \vdots \\ 1_{T,1} & \cdots & 1_{T,T} \end{pmatrix} = \begin{pmatrix} T_{1,1} & \cdots & T_{1,T} \\ \vdots & \ddots & \vdots \\ T_{T,1} & \cdots & T_{T,T} \end{pmatrix} = Tii'$$

Substitute $ii'ii'$ with Tii' in (1) and we can obtain:

$$(I_T - \frac{1}{T}ii')(I_T - \frac{1}{T}ii') = I_T - \frac{1}{T}ii'$$

This proves that $I_T - \frac{1}{T}ii'$ is a idempotent matrix

2 Question 2

2.1 (6)

Expand the numerator we can obtain:

$$\frac{(X - ui)'(X - ui)}{\sigma^2} = \frac{\sum_{i=1}^n (X - u)^2}{\sigma^2} = \sum_{i=1}^n \left(\frac{X - u}{\sigma}\right)^2$$

Obviously $\frac{X - u}{\sigma} \sim N(0, 1)$, thus the above expression is the sum of nth squared standard normal distribution. i.e.

$$\frac{(X - ui)'(X - ui)}{\sigma^2} \sim \chi^2(n)$$

2.2 (7)

In the previous Q1(4) we have proved that:

$$(I_n - \frac{1}{n}ii')X = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} - \begin{pmatrix} \bar{x} \\ \bar{x} \\ \vdots \\ \bar{x} \end{pmatrix} = X - \bar{X}$$

In the same manner:

$$(I_n - \frac{1}{n}ii')ui = u(I_n - \frac{1}{n}ii')i = u(i - \bar{i}) = 0$$

where u is a scalar and $i = \bar{i} = (1, \dots, 1)'$

Next applying this result to $(I_n - \frac{1}{n}ii')(X - ui)$, we can obtain:

$$(I_n - \frac{1}{n}ii')(X - ui) = (I_n - \frac{1}{n}ii')X - (I_n - \frac{1}{n}ii')ui = (X - \bar{X})$$

Thus we proved that:

$$\begin{aligned}(X - ui)'(I_n - \frac{1}{n}ii')(X - ui) &= (X - ui)'(I_n - \frac{1}{n}ii')(I_n - \frac{1}{n}ii')(X - ui) \\ &= (X - \bar{X})'(X - \bar{X}) = \sum_{i=1}^n (X_i - \bar{X})^2\end{aligned}$$

2.3 (8)

Applying the same method in Q1(1), we can obtain:

$$\frac{(X - ui)'(I_n - \frac{1}{n}ii')(X - ui)}{\sigma^2} \sim \chi^2(\text{tr}(I_n - \frac{1}{n}ii'))$$

where $\text{tr}(I_n - \frac{1}{n}ii') = \text{tr}(I_n) - \text{tr}(\frac{1}{n}ii') = n - 1$

Thus:

$$\frac{(X - ui)'(I_n - \frac{1}{n}ii')(X - ui)}{\sigma^2} \sim \chi^2(n - 1)$$